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Retrieval of vertical T , H_2O , O_3 , and CO_2 profiles, with sensitivity analysis (SA) and uncertainty quantification (UQ)

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Objectives



- **Augment retrieval methods for temperature (T), H_2O , O_3 , and CO_2 to enable vertical-profile estimation for CO_2 , with embedded sensitivity analysis (SA) and uncertainty quantification (UQ)**
 - Adapt and evolve existing algorithmic methods and tools to support sensitivity analysis (SA) and uncertainty quantification (UQ)
 - Embed in analyses of maximum-likelihood estimation (MLE) retrievals to support and improve:
 - instrument calibration,
 - algorithm verification, and
 - data validation.
- **Assess differences between retrieval approaches, e.g., “Optimal Estimation (OE)” and “Vanishing Partial Derivatives (VPD)”, and map into a single unified scheme**
- **Characterize and quantify retrieval errors. Use results to:**
 - identify dominant sensitivities to improve retrieval algorithms, and
 - assess and qualify data variances between retrievals and validation
- **Explore application to vertical-profile estimations, extending to near-surface retrievals**



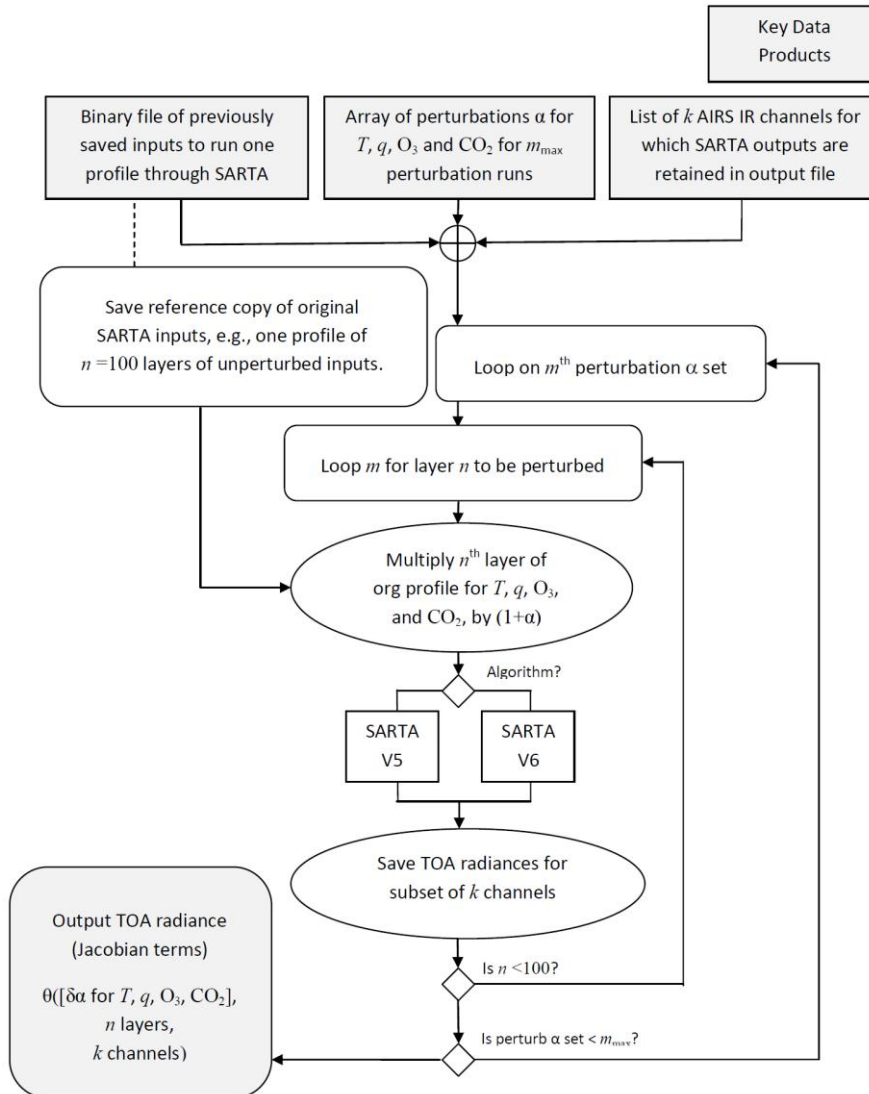
Approach



- **Formulation of sensitivity analysis (SA) and uncertainty quantification (UQ) framework and methods**
 - Adopt maximum-likelihood estimation (MLE) methods
 - Develop concepts for Monte Carlo methods, DAKOTA,* linearity tests, ...
 - Explore projection onto vertical-profile basis functions.
- **Extend/adapt existing analysis tools, and develop new tools, as necessary, to support SA/UQ functionality**
 - Tools that employ the Standalone AIRS Radiative-Transfer Algorithm (SARTA) for the analysis of channel selection and the calculation of retrieval Jacobians, in support of SA/UQ
 - Adapt contribution-function estimation for vertical-profile retrieval data
- **Adapt observation-error analyses, based on maximum-likelihood estimation (MLE) methods**
 - Link to analyses of frequency-calibration shifts.
 - Examine explicit and implicit assumptions in MLE estimations and compare “OE” and “VPD” approaches
 - Perform preliminary/idealized UQ analyses to explore and demonstrate UQ concepts and formulations.

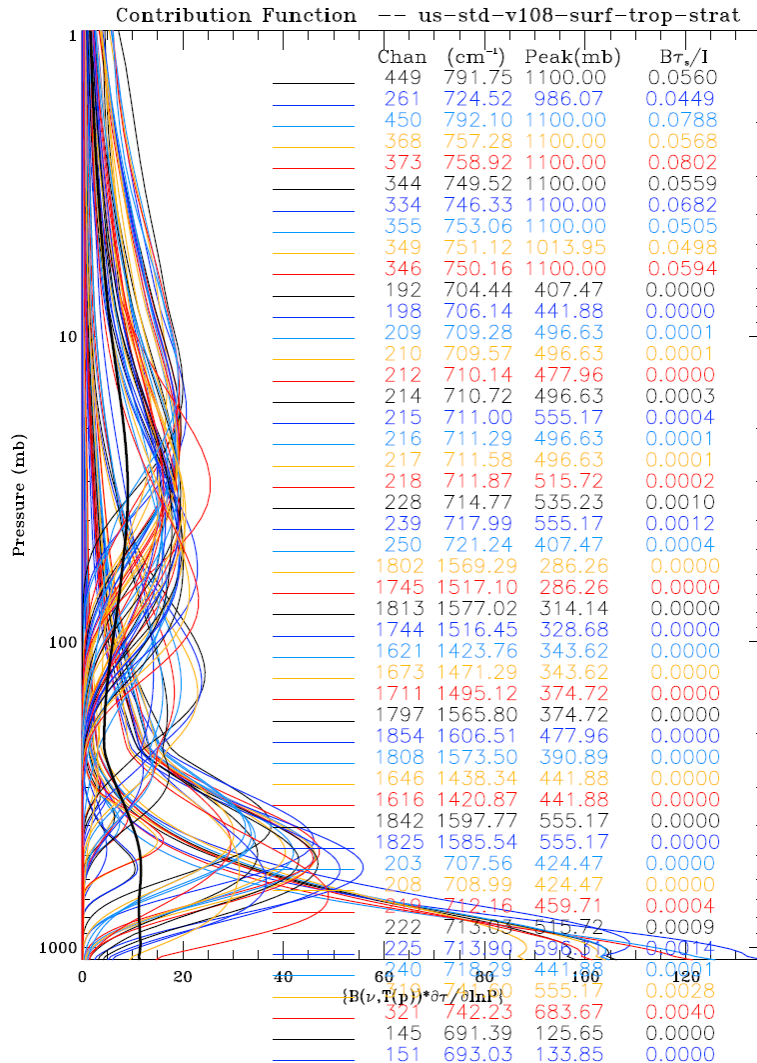


AIRS profile-analysis tool



- **Definitions:**
 - TOA: Top of atmosphere
 - SARTA: Standalone AIRS Radiative-Transfer Algorithm
- **Solution for $n_{\max}=100$ vertical ($\ln p$) layers**
- **In minimizing cost functions for data retrieval, iterations allow the i^{th} input vertical profile, $x_i(\ln p)$, to be perturbed by a factor $(1+\alpha_i)$**
 - At present, the α_i are assumed uniform in altitude ($\ln p$).
 - $i = T, q, O_3, \text{ and } CO_2$

CO₂ channel contribution functions



Radiance intensity*

$$I_{\nu} = \underbrace{S(\nu, T_0, \varepsilon_{\nu})}_{\text{Surface contribution}} - \underbrace{\int_0^{p_0} B[\nu, T(p)] \frac{\partial \tau(\nu, p, \dots)}{\partial p} dp}_{\text{Weighting function (WF)}}$$

Contribution function (CF)

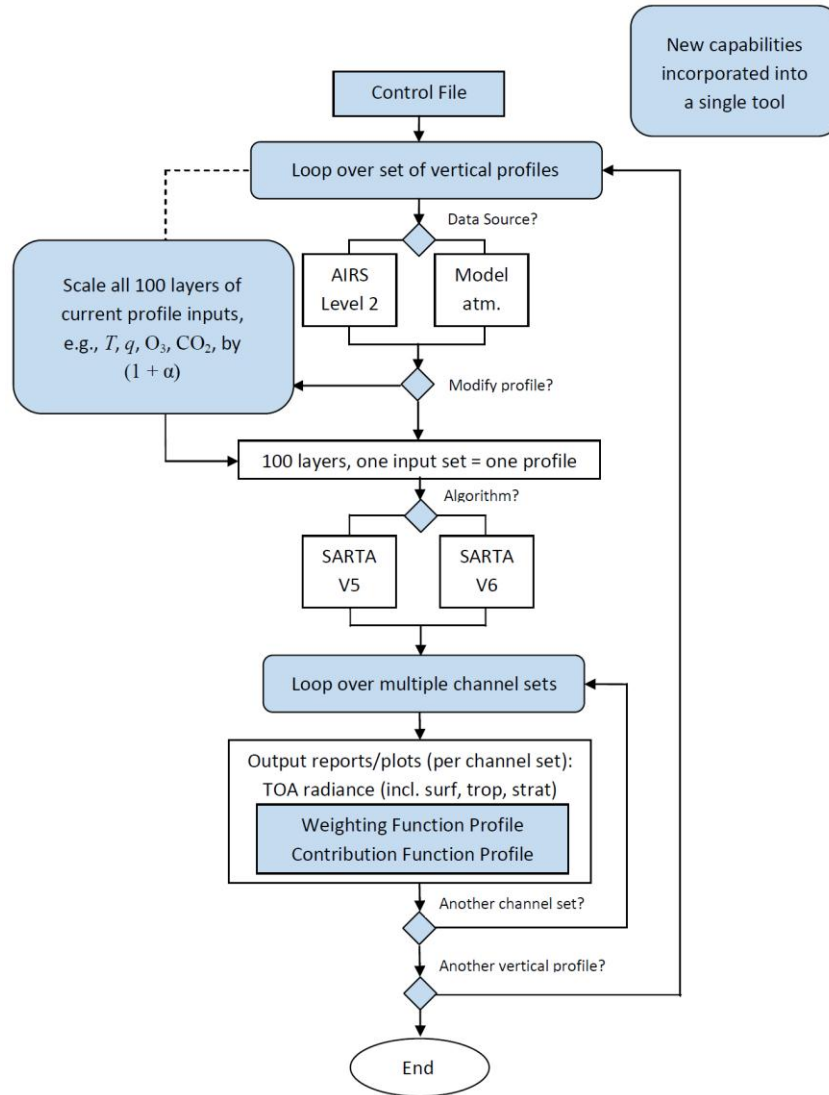
CF plots of channels previously identified as relevant for CO₂ retrievals in the

- stratosphere,
- troposphere, and
- near-surface

Plots estimated for U.S. std. atm.

- Real atmosphere varies with place and time (location, altitude, weather, season, composition, etc.)

AIRS weighting-/contribution-function tool



- **Definitions:**
 - TOA: Top of atmosphere
 - SARTA: Standalone AIRS Radiative-Transfer Algorithm
- **For each channel, the tool estimates the layer profiles for the**
 - weighting function (WF)
 - contribution function (CF)
- **Scales all profiles using uniform $(1 + \alpha_i)$ scaling factors from the AIRS profile-analysis tool**
 - Solution for $n_{\max} = 100$ vertical ($\ln p$) layers



Updated retrieval-tool capabilities (at present)



- **Can ingest vertical columns of properties/constituents (T_{air} , H_2O , O_3 , CO_2 , and others) to SARTA and can compute:**
 - Top-of-atmosphere (TOA) radiance for the AIRS 2378 IR channel set, including contributions from surface, troposphere, and stratosphere.
 - Weighting function (WF): layer-wise calculation of $d\tau(\nu, \ln p, \dots)/d\ln(p)$
 - Contribution function (CF): $B[\nu, T(p)] d\tau(\nu, \ln p, \dots)/d\ln(p)$, net TOA radiance,
 - per channel k ,
 - per level n ,
 - for a given temperature profile, $T_{\text{air}}(p)$
 - Analysis tools to support processing of multiple permutations of a single profile
 - Custom software wrappers to enable running retrieval tools for SA/UQ
- **Enable perturbation of vertical constituent column inputs by fixed scale factors, $\alpha_i(p)$, simultaneously for all levels**
 - Perturbation, layer-by-layer, and collation of changes in radiance, in response to changes in inputs, to support Jacobian/SA studies

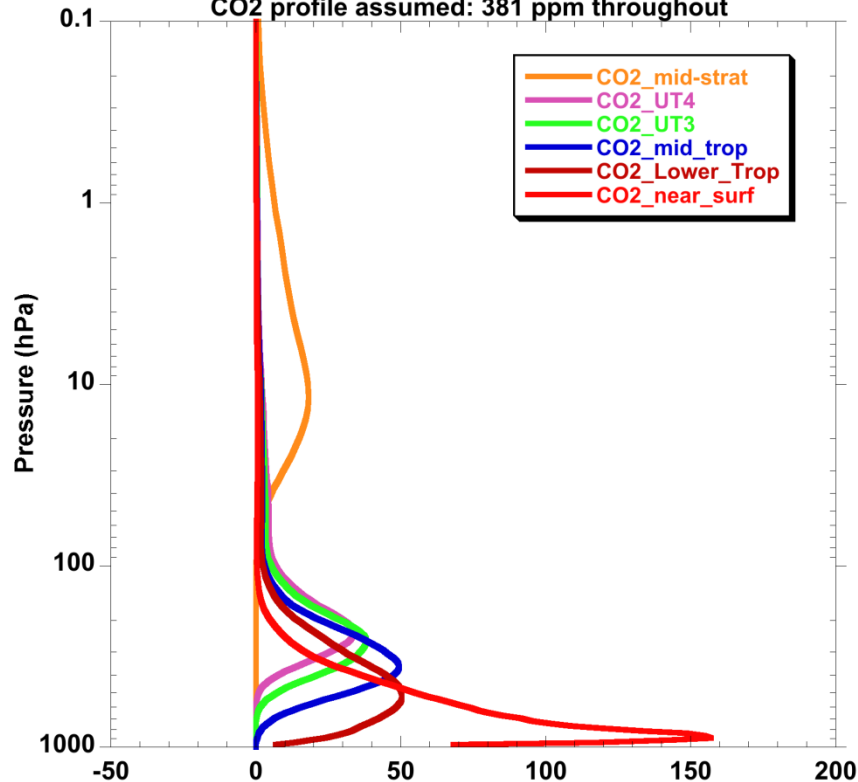


Examples of CF aggregates for CO₂

AIRS Level 2 Footprint Tropical Pacific Atmospheric Profiles
1 Jan 2007; Granule 231; FP01; SC02

Lat: 00.46S Lon: 149.29W

CO₂ profile assumed: 381 ppm throughout

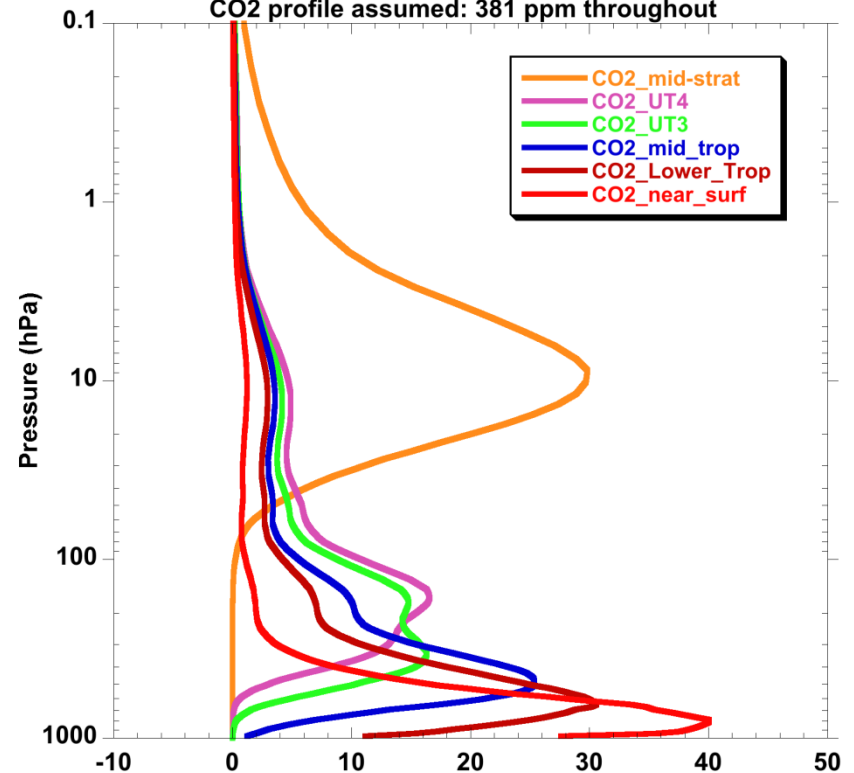


Average Contribution Function, $B(T) \times \Delta\tau / \Delta \ln(p)$

AIRS Level 2 Footprint Frozen Arctic Ocean Atmospheric Profiles
1 Jan 2007; Granule 202; FP01; SC02

Lat: 74.86N Lon: 178.39W

CO₂ profile assumed: 381 ppm throughout



Average Contribution Function, $B(T) \times \Delta\tau / \Delta \ln(p)$

- Sufficient variance between CFs with respect to p to allow CO₂ (and other) vertical-profile estimations



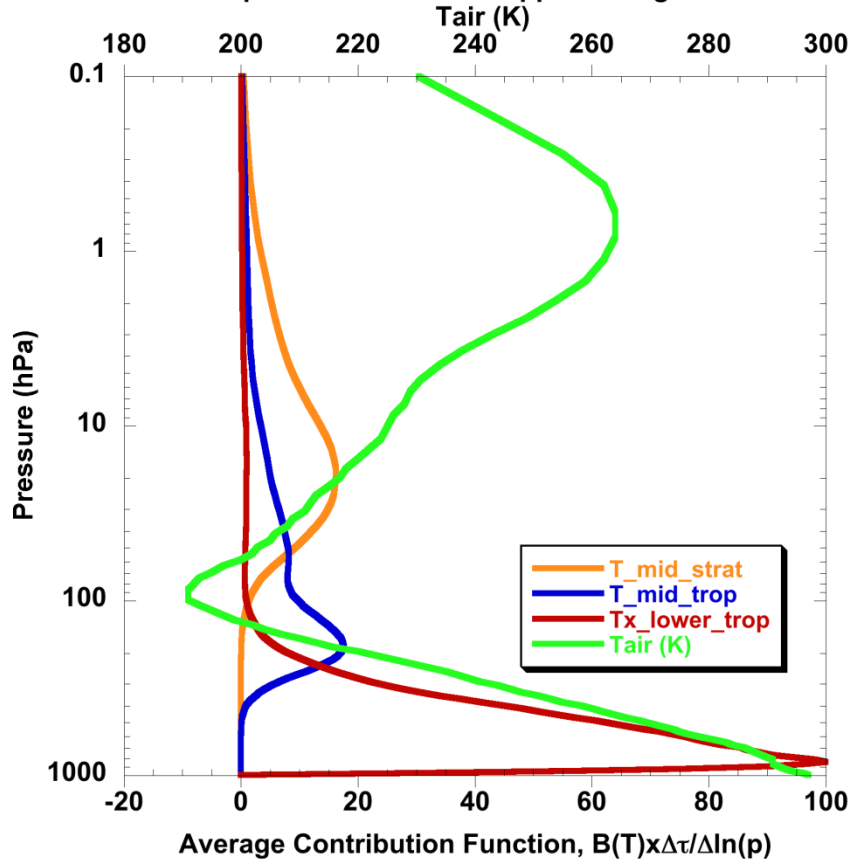
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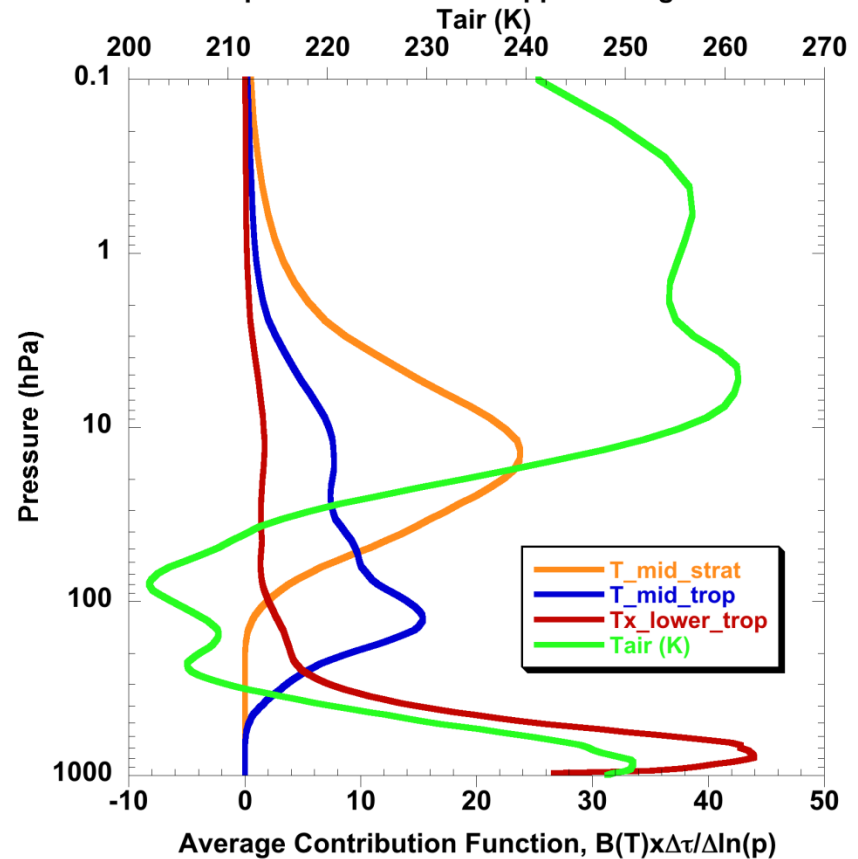


Select CF aggregates with T profiles

AIRS Level 2 Footprint Tropical Pacific Atmospheric Profiles
1 Jan 2007; Granule 231; FP01; SC02
Lat: 00.46S Lon: 149.29W
CO₂ profile assumed: 381 ppm throughout



AIRS Level 2 Footprint Frozen Arctic Ocean Atmospheric Profiles
1 Jan 2007; Granule 202; FP01; SC02
Lat: 74.86N Lon: 178.39W
CO₂ profile assumed: 381 ppm throughout



- Present algorithms/tools allow for small, but uniform, multiplicative shifts $(1+\alpha_i)$ for i^{th} profile: adequate for retrievals over narrow p ranges.



Maximum Likelihood Estimation (MLE) with UQ

Simple example



TOA radiances calculation (SARTA output)

$\theta_{v_m} = \theta_{v_m}(x_T, x_q, x_{O_3}, \dots, x_{CO_2}); x_i = x_i(n), n = 1, \dots, n_{\max} = 100$, is the i^{th} variable column vector

Maximize conditional probability*

$$p(x_{CO_2} | \theta_{v_1}^{\text{obs}}, \theta_{v_2}^{\text{obs}}, \dots, \theta_{v_M}^{\text{obs}}; x_T^b, x_q^b, x_{O_3}^b, x_{CO_2}^b, \dots)$$

Minimize the cost function (varying x_{CO_2} only)

$$J(x_{CO_2}) = \frac{1}{2} \sum_{m=1}^M [\theta_{v_m}^{\text{obs}} - \theta_{v_m}(x_T^b, x_q^b, x_{O_3}^b, x_{CO_2}^b)]^T R_{v_m v_m}^{-1} [\theta_{v_m}^{\text{obs}} - \theta_{v_m}(x_T^b, x_q^b, x_{O_3}^b, x_{CO_2}^b)] \\ + \frac{1}{2} (x_{CO_2} - x_{CO_2}^b)^T B_{CO_2}^{-1} (x_{CO_2} - x_{CO_2}^b)$$

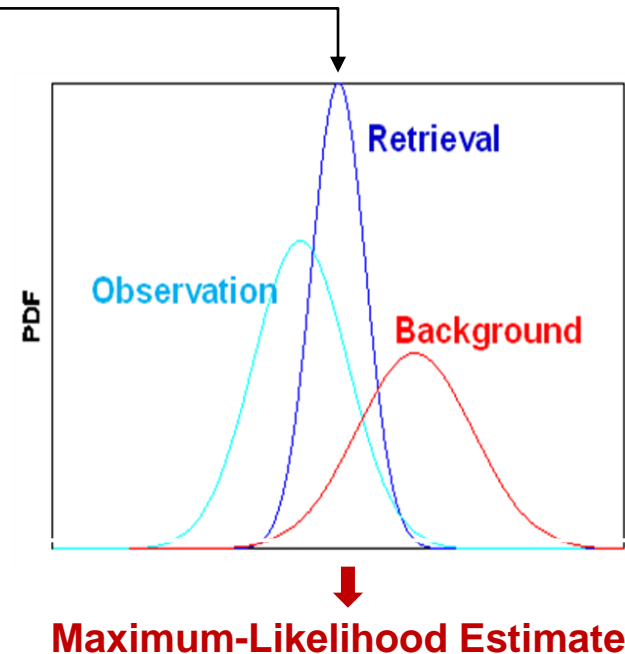
Retrieved CO_2 error covariance

$$P_{CO_2} = (B^{-1} + H_{CO_2}^T R^{-1} H_{CO_2})^{-1} \\ = B - B H_{CO_2}^T (H_{CO_2} B H_{CO_2}^T + R)^{-1} H_{CO_2} B$$

B: Background-error covariance

R: Observational-error covariance

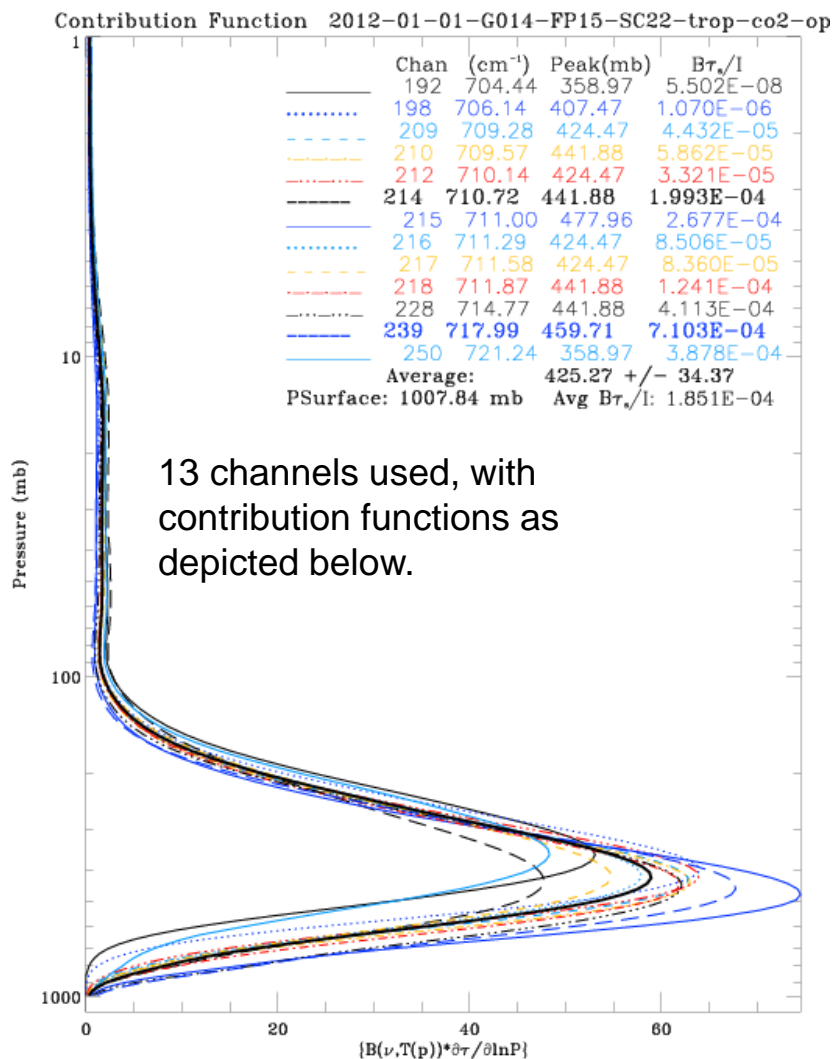
H_{CO₂}: Jacobian, sensitivity matrix



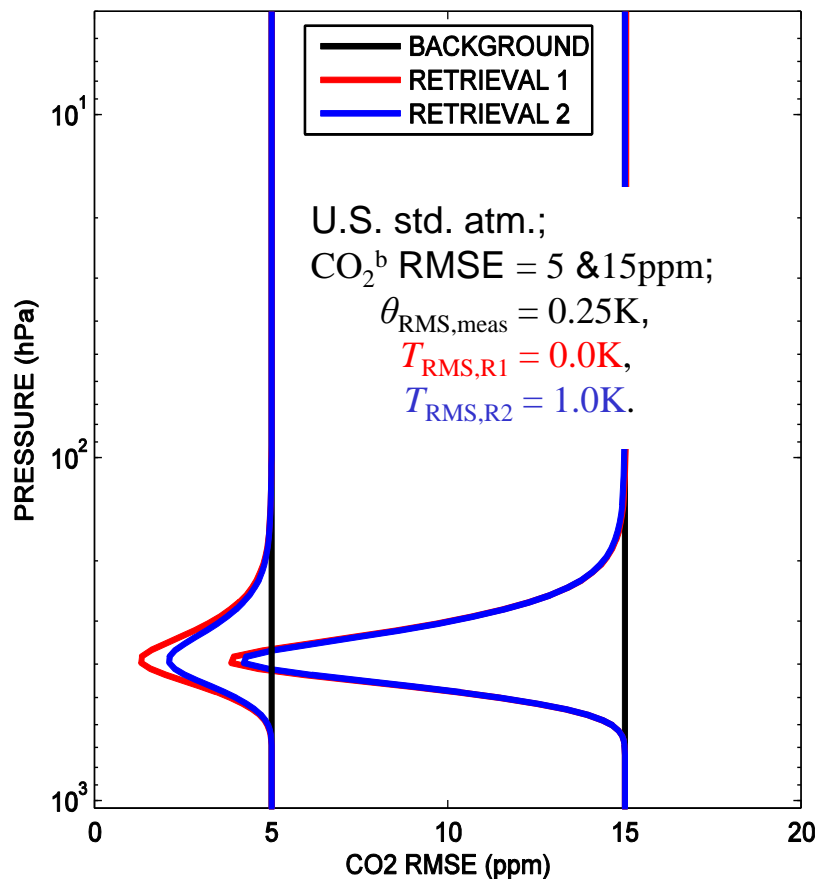
* Superscripts: "obs"=observation/measurement,
"b"=background/priors



A simple illustration of retrieved CO₂ RMSE



Dependence of retrieved CO₂ RMSE on background T and CO₂ (priors) RMSE.





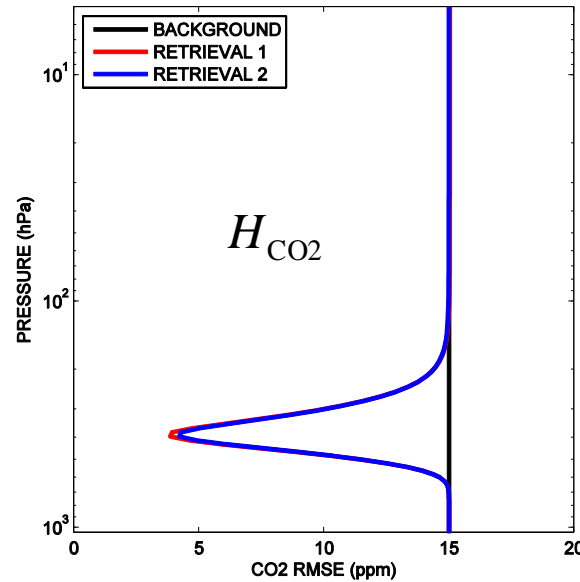
Channel brightness temperatures and their sensitivity to CO₂, and retrieved CO₂ RMSE



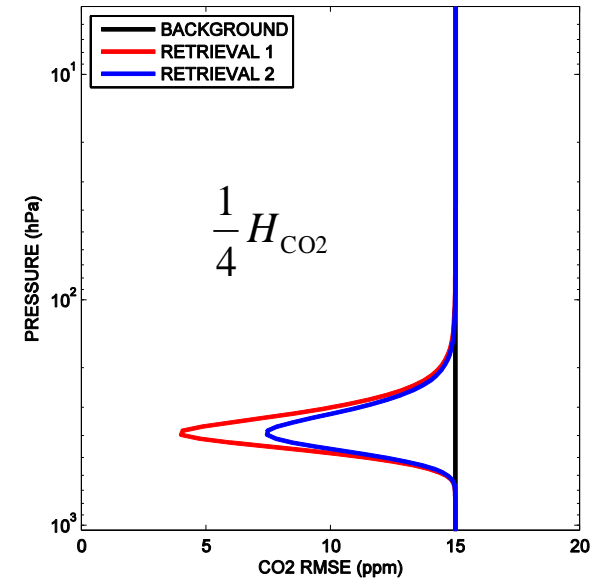
Jacobian/sensitivity*

$$H_{\text{CO}_2} = \left[\begin{array}{c} \frac{\partial \theta_{v_1}(x_T^b, x_q^b, x_{\text{O}_3}^b, x_{\text{CO}_2}^b)}{\partial x_{\text{CO}_2}^t}, \\ \frac{\partial \theta_{v_2}(x_T^b, x_q^b, x_{\text{O}_3}^b, x_{\text{CO}_2}^b)}{\partial x_{\text{CO}_2}^t}, \\ \dots, \\ \frac{\partial \theta_{v_M}(x_T^b, x_q^b, x_{\text{O}_3}^b, x_{\text{CO}_2}^b)}{\partial x_{\text{CO}_2}^t} \end{array} \right]$$

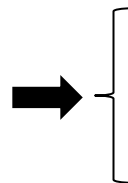
High sensitivities



Lower sensitivities



Importance of
high sensitivities



1. Lower retrieval RMSE
2. Less-sensitive to errors in other parameters
3. Channel selection and retrieval algorithms



Future plans



- **Further extend and generalize SARTA use**
 - Support systematic, large-scale sensitivity analyses (SA)
 - Allow for p -dependent input scaling functions, $\alpha_i(\ln p)$.
 - Explore and support Monte Carlo techniques and DAKOTA for retrieval UQ
- **Perform UQ on multi-season mid-tropospheric retrievals**
 - Compare with V5 retrievals
 - Explore relation and application to V6 retrievals
- **Extend SA/UQ concepts and formulations and assess their applicability to the retrieval of vertical profiles for:**
 - stratospheric layers,
 - low-troposphere, and
 - near-surface retrievals.
- **Explore functional-projection analyses for vertical-profile estimation**
- **Apply to instrument calibration, enhance algorithm verification, and use available data for quantitative validation**



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Thank you.